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Multi-particle azimuthal correlations for extracting event-by-event elliptic and triangular flow in Au+Au collisions at $\sqrt{s_{NN}} = 200$ GeV


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We present measurements of elliptic and triangular azimuthal anisotropy of charged particles detected at forward rapidity $1 < |\eta| < 3$ in Au+Au collisions at $\sqrt{s_{NN}} = 200$ GeV, as a function of centrality. The multiparticle cumulant technique is used to obtain the elliptic flow coefficients $v_2\{2\}$, $v_2\{4\}$, $v_3\{6\}$, and $v_2\{8\}$, and triangular flow coefficients $v_3\{2\}$ and $v_3\{4\}$. Using the small-variance limit, we estimate the mean and variance of the event-by-event $v_2$ distribution from $v_2\{2\}$ and $v_2\{4\}$. In a complementary analysis, we also use a folding procedure to study the distributions of $v_2$ and $v_3$ directly, extracting both the mean and variance. Implications for initial geometrical fluctuations and their translation into the final state momentum distributions are discussed.

I. INTRODUCTION

Collisions of heavy nuclei at ultra-relativistic energies are believed to create a state of matter called the strongly coupled quark-gluon plasma, as first observed at the Relativistic Heavy Ion Collider (RHIC) \cite{1-4}. The quark-gluon plasma evolves hydrodynamically as a nearly perfect liquid as evinced by the wealth of experimental measurements and theoretical predictions (or descriptions) of the azimuthal anisotropy of the produced particles. \cite{5}. Multi-particle correlations are generally taken as strong evidence of hydrodynamical flow, which necessarily affects most or all particles in the event \cite{6}. This is different from mimic correlations (generically called nonflow) that are not related to the hydrodynamical evolution and typically involve only a few particles.

Multi-particle correlations are also interesting because they have different sensitivities to the underlying event-by-event fluctuations, which can provide additional insights into the initial geometry and its translation into final state particle distributions \cite{7}\cite{8}.

Recently, experimental and theoretical efforts have been directed towards measuring the fluctuations directly, using event-by-event unfolding techniques. In principle, the multi-particle correlations and unfolding techniques provide the same information about the underlying fluctuations, though in practice with different sensitivities \cite{9}. The techniques used at the Large Hadron Collider (LHC) are experimentally very different and provide complementary information \cite{10}\cite{11}.

In this manuscript we present measurements of 2-, 4-, 6-, and 8-particle correlations as well as event-by-event measurements of the azimuthal anisotropy parameters corresponding to elliptic $v_2$ and triangular $v_3$ flow. We estimate the relationship between the mean and variance with both techniques and discuss the implications for understanding the detailed shape of the $v_2$ and $v_3$ distributions. These measurements, while the first of their kind at forward rapidity, are consistent with previous measurements at midrapidity by STAR \cite{12} and PHOBOS \cite{13}.

II. EXPERIMENTAL SETUP

In 2014, the PHENIX experiment \cite{14} at RHIC collected nearly $2 \times 10^{10}$ minimum bias (MB) events of Au+Au collisions at a nucleon-nucleon center-of-mass energy $\sqrt{s_{NN}} = 200$ GeV. The present analysis makes use of a subset ($\approx 10^9$ events) of the total 2014 data sample. The PHENIX beam-beam counters (BBC) are used for triggering and centrality determination. The BBCs \cite{15} are located $\pm 144$ cm from the nominal interaction point and cover the full azimuth and $3.1 < |\eta| < 3.9$ in pseudorapidity. By convention, the north side is forward rapidity ($\eta > 0$) and the south side is backward rapidity ($\eta < 0$). Each BBC comprises an array of 64 phototubes with a fused quartz Čerenkov radiator on the front. Charged particles impinging on the radiator produce Čerenkov light which is then amplified and detected by the phototube. The PHENIX MB trigger for the 2014 data sample of Au+Au collisions at $\sqrt{s_{NN}} = 200$ GeV was defined by at least two phototubes in each side of the BBC having signal above threshold and an online z-vertex within $\pm 10$ cm of the nominal interaction point. Additionally, PHENIX has a set of zero-degree calorimeters (ZDC) that measure spectator neutrons from each incoming nucleus \cite{15}. We require a minimum energy in both ZDCs to remove beam related background present at the highest luminosities.

The centrality definition is based on the combined signal in the north and south BBCs. The charge distribution is fitted using a Monte Carlo (MC) Glauber \cite{16} simulation to estimate the number of participating nucleons ($N_{\text{part}}$) and a negative binomial distribution to describe the BBC signal for fixed $N_{\text{part}}$. All quantities in the present manuscript are reported as a function of centrality and the corresponding $N_{\text{part}}$ values are shown in Table \ref{table:1}.

The main detector used in the analysis is the forward silicon vertex detector (FVTX). The FVTX \cite{17} is a silicon strip detector comprising two arms, north and south, covering $1 < |\eta| < 3$. In Au+Au collisions there is a strong correlation between the total signal in the BBCs and the total number of tracks in the FVTX. To remove beam related background, we apply an additional event selection on the correlation between the total BBC signal and the number of tracks in the FVTX.

Each FVTX arm has four layers. In the track reconstruction software, a minimum of three hits is required to reconstruct a track. However, it is possible for there to be hit sharing with the central rapidity detector (VTX),

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TABLE I. $N_{\text{part}}$ values for various centrality categories.

<table>
<thead>
<tr>
<th>Centrality</th>
<th>$\langle N_{\text{part}} \rangle$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0%–5%</td>
<td>$350.8 \pm 3.1$</td>
</tr>
<tr>
<td>5%–10%</td>
<td>$301.7 \pm 4.7$</td>
</tr>
<tr>
<td>10%–20%</td>
<td>$236.1 \pm 3.8$</td>
</tr>
<tr>
<td>20%–30%</td>
<td>$167.6 \pm 5.5$</td>
</tr>
<tr>
<td>30%–40%</td>
<td>$115.5 \pm 5.8$</td>
</tr>
<tr>
<td>40%–50%</td>
<td>$76.1 \pm 5.5$</td>
</tr>
<tr>
<td>50%–60%</td>
<td>$47.0 \pm 4.7$</td>
</tr>
<tr>
<td>60%–70%</td>
<td>$26.7 \pm 3.6$</td>
</tr>
<tr>
<td>70%–80%</td>
<td>$13.6 \pm 2.4$</td>
</tr>
<tr>
<td>80%–93%</td>
<td>$6.1 \pm 1.3$</td>
</tr>
</tbody>
</table>

so that one or two of the three required hits can be in the VTX. We select tracks using a stricter requirement of at least three hits in FVTX, irrespective of the number of hits in the VTX. We further require that the track reconstruction algorithm have a goodness of fit of $\chi^2$/d.o.f. < 5 for each track. Lastly, we require that each track has a distance of closest approach (DCA) of less than 2 cm. The DCA is defined as the distance between the event vertex and the straight-line extrapolation point of the FVTX track onto a plane which is perpendicular to the $z$-axis and contains the event vertex. A 2 cm cut selects the FVTX tracks that likely originate from the event vertex, and is conservative in accepting the nonzero DCA tail that stems from the uncertainty in the determination of the vertex position and the bending of the actual track in the experimental magnetic field. Due to the orientation of the FVTX strips relative to the magnetic field, momentum determination is not possible using the tracks in the FVTX alone. However, GEANT-4 simulations have determined that the tracking efficiency is relatively independent of momentum for $p_T \gtrsim 0.3$ GeV/c. Figure 1 shows the $p_T$ dependence of the FVTX tracking efficiency averaged over $1 < |\eta| < 3$. Figure 2 shows the tracking efficiency as a function of $\eta$ in the FVTX for two different $z$-vertex selections. The single particle tracking efficiency has a maximum value of 98.6% as a function of $\eta$. When averaging over $1 < |\eta| < 3$, the maximum value of the $p_T$-dependent efficiency is 67.9%, and $p_T = 0.3$ GeV/c the efficiency is at 75% of its maximum value.

### III. ANALYSIS METHODS

The azimuthal distribution of particles in an event can be represented by a Fourier series [19]:

$$\frac{dN}{d\phi} \propto 1 + \sum_n 2v_n \cos(n(\phi - \psi_n)), \quad (1)$$

where $n$ is the harmonic number, $\phi$ is the azimuthal angle of some particle, $\psi_n$ is the symmetry plane, and $v_n = \langle \cos(n(\phi - \psi_n)) \rangle$. There are many experimental techniques for estimating the $v_n$ coefficients, some of which we discuss in this section.

The main ingredient in the present analysis, for both the cumulant results and the folding results, is the Q-vector. The Q-vector is a complex number $Q_n = Q_{n,x} +$
iQ_{n,y} with the components defined as

\begin{align}
\Re Q_n &= Q_{n,x} = \sum_{i}^{N} \cos n\phi_i, \\
\Im Q_n &= Q_{n,y} = \sum_{i}^{N} \sin n\phi_i,
\end{align}

where \(\phi_i\) is the azimuthal angle of some particle and \(N\) is the number of particles in some event or subevent—a subevent is a subset of a whole event, usually selected based on some kinematic selection, e.g. pseudorapidity. Because the PHENIX FVTX detector subsystem is split into two separate arms, north \((1 < \eta < 3)\) and south \((-3 < \eta < -1)\), it is natural to use tracks in the two arms as separate subevents for some calculations. In other calculations, all tracks from north and south will be combined into a single event.

Additional corrections to the data are needed to account for any nonuniformity in the azimuthal acceptance. In the case of uniform azimuthal acceptance, the event average of the Q-vector components is zero: \(\langle Q_{n,x} \rangle = \langle Q_{n,y} \rangle = 0\). In the case of nonuniform acceptance, there can be a systematic bias such that this relation does not hold. In the case of few-particle correlations, i.e. 2- and 4-particle correlations, the bias can be corrected analytically in a straightforward manner \[20\]. In the case of correlations with a larger number of particles, however, this becomes impractical. The total number of terms in a k-particle cumulant calculation without the assumption of azimuthal uniformity is given by the Bell sequence: 1, 2, 5, 15, 52, 203, 877, 4140, ...—that is, the number of terms for 2- and 4-particle correlations is a rather manageable 2 and 15, respectively; contrariwise, the number of terms for 6- and 8-particle correlations is a rather unmanageable 203 and 4140 terms, respectively. For that reason, the only practicable choice is to perform calculations on corrected Q-vectors. The present analysis makes use of Q-vector re-centering \[21\]. In this procedure one has the relation

\[Q_{\text{corrected}} = Q_{\text{raw}} - Q_{\text{average}},\]

where

\[Q_{\text{average}} = N\langle \cos n\phi \rangle + iN\langle \sin n\phi \rangle,\]

and

\[\langle \cos n\phi \rangle = \langle Q_{n,x} \rangle / N,\]
\[\langle \sin n\phi \rangle = \langle Q_{n,y} \rangle / N.\]

In the present analysis, we perform the Q-vector re-centering procedure for each FVTX arm separately and as a function of \(N_{\text{FVTX tracks}}\). To assess the associated systematic uncertainty, we perform the Q-vector re-centering as a function of centrality instead, as a function of additional secondary variables (event vertex and operational time period), and for combined arms instead of separate.

### A. Cumulants

The cumulant method for flow analysis was first proposed in Ref. \[22\]. In the present analysis, we use the recursion algorithm developed in Ref. \[23\], which is a generalization of the direct calculations using Q-vector algebra first derived in Ref. \[20\]. We consider 2-, 4-, 6-, and 8-particle correlations. The multi-particle correlations are denoted \(\langle k \rangle\) for \(k\)-particle correlations and are estimators for the \(k\)-th moment of \(v_n\), i.e. \(\langle k \rangle = \langle v_n^k \rangle\). In terms of the angular relationships between different particles, they are

\[\langle 2 \rangle = \langle \cos(n(\phi_1 - \phi_2)) \rangle,\]
\[\langle 4 \rangle = \langle \cos(n(\phi_1 + \phi_2 - \phi_3 - \phi_4)) \rangle,\]
\[\langle 6 \rangle = \langle \cos(n(\phi_1 + \phi_2 + \phi_3 - \phi_4 - \phi_5 - \phi_6)) \rangle,\]
\[\langle 8 \rangle = \langle \cos(n(\phi_1 + \phi_2 + \phi_3 + \phi_4 - \phi_5 - \phi_6 - \phi_7 - \phi_8)) \rangle,\]

where \(\phi_1, ..., \phi_8\) represent the azimuthal angles of different particles in the event.

The \(k\)-particle cumulants, denoted \(c_n \{k\}\) are constructed in such a way that potential contributions from lower order correlations are removed. Because the cumulants mix various terms that are of equal powers of \(v_n\), the cumulant method \(v_n\), denoted \(v_n \{k\}\), is proportional to the \(k\)-th root of the cumulant. The \(c_n \{k\}\) are constructed as follows:

\[c_n \{2\} = \langle 2 \rangle,\]
\[c_n \{4\} = \langle 4 \rangle - 2\langle 2 \rangle^2,\]
\[c_n \{6\} = \langle 6 \rangle - 9\langle 4\rangle \langle 2 \rangle + 12\langle 2 \rangle^3,\]
\[c_n \{8\} = \langle 8 \rangle - 16\langle 6\rangle \langle 2 \rangle - 18\langle 4 \rangle^2 + 144\langle 4 \rangle \langle 2 \rangle^2 - 144\langle 2 \rangle^4,\]

and the \(v_n \{k\}\) are

\[v_n \{2\} = (c_n \{2\})^{1/2},\]
\[v_n \{4\} = (-c_n \{4\})^{1/4},\]
\[v_n \{6\} = (c_n \{6\})^{1/6},\]
\[v_n \{8\} = (-c_n \{8\})^{1/8}.\]

It is also possible to construct cumulants in two or more subevents, though in the present analysis we will only concern ourselves with two subevents. For 2-particle correlations, rather than \(\langle 2 \rangle = \langle \cos(n(\phi_1 - \phi_2)) \rangle\) where \(\phi_1\) and \(\phi_2\) are from the same subevent, one has instead \(\langle 2 \rangle_{\text{sub}} = \langle \cos(n\phi_{a1} - \phi_{a2}) \rangle\) where \(a, b\) denote two different subevents. The cumulant and \(v_n\) have the same relationship as in the single event case, i.e. \(v_n \{2\}_{\text{sub}} = c_n \{2\}_{\text{sub}} = \langle 2 \rangle_{\text{sub}}\). The two-subevent 2-particle cumulant is also known as the scalar product method \[24\].

Subevent cumulants for correlations with four or more particles were first proposed in Ref. \[25\]. For two-subevent 4-particle correlations, there are two possibil-
where the former allows 2-particle correlations within single subevents and the latter excludes them. The latter is therefore less susceptible to nonflow than the former, although both are less susceptible to nonflow than single event 4-particle correlations. The cumulants take the form

\begin{align}
\langle 4 \rangle_{ab|ab} &= \langle 4 \rangle_{ab|ab} - \langle 2 \rangle_{a|a} \langle 2 \rangle_{b|b} - \langle 2 \rangle_{a|b}^2, \\
\langle 4 \rangle_{aa|bb} &= \langle 4 \rangle_{aa|bb} - \langle 2 \rangle_{a|a}^2,
\end{align}

and the \( v_n \{4\} \) values have the same relationship to the cumulants as in the single particle case, i.e. \( v_n \{4\}_{ab|ab} = (-c_n \{4\}_{ab|ab})^{1/4} \) and \( v_n \{4\}_{aa|bb} = (c_n \{4\}_{aa|bb})^{1/4} \).

To determine systematic uncertainties associated with event and track selection for the cumulant analysis, we vary the event and track selection criteria and assess the variation on the final analysis results. The z-vertex selection is modified from \( (\eta < \eta_i) \) to \( (\eta < \eta_i < \eta_f) \)

For the unfolding, ATLAS determines the response matrix in a data driven way. The smearing in the response matrix is modest as Pb+Pb collisions have a high multiplicity and the ATLAS detector has large phase space coverage for tracks \(-2.5 < \eta < 2.5\). In our case, the multiplicity of Au+Au collisions is lower in comparison with the multiplicity in Pb+Pb collisions and the phase space coverage of the FVTX detector is significantly smaller. Hence, the smearing as encoded in the response matrix is significantly larger and the unfolding is more challenging.

To estimate the response function of the detector, we follow the procedure from ATLAS \[10\], which is to examine the difference between two subevents for both \( Q_x \) and \( Q_y \). We compare the \( Q \)-vector determined in the south arm of the FVTX, \( Q_{south} \), to the \( Q \)-vector determined in the north arm of the FVTX, \( Q_{north} \). Figure 3 shows an example of this procedure for the \( n = 2 \) case. Figure 3(a) shows the 2-dimensional distribution for the 20\%–30\% centrality selection; Fig 3(b) shows the one-dimensional projection of this onto the \( x \)-axis (i.e. the one-dimensional distribution of \( Q_{north} - Q_{south} \)); Fig 3(c) shows the one-dimensional projection of this onto the \( y \)-axis (i.e. the one-dimensional distribution of \( Q_{north} - Q_{south} \)). These distributions in all centrality selections are Gaussian over four orders of magnitude, and we characterize them via their Gaussian widths \( \delta_{2SE} \) which are given in Table III. It is notable that these widths are more than a factor of two larger than those quoted by ATLAS in Pb+Pb collisions, for example \( \delta_{2SE} = 0.050 \) for Pb+Pb 20\%–25\% central events \[10\].

<table>
<thead>
<tr>
<th>Centrality</th>
<th>( \delta_{2SE}(n = 2) )</th>
<th>( \delta_{2SE}(n = 3) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>0%–5%</td>
<td>0.117</td>
<td>0.115</td>
</tr>
<tr>
<td>5%–10%</td>
<td>0.115</td>
<td>0.113</td>
</tr>
<tr>
<td>10%–20%</td>
<td>0.115</td>
<td>0.113</td>
</tr>
<tr>
<td>20%–30%</td>
<td>0.121</td>
<td>0.119</td>
</tr>
<tr>
<td>30%–40%</td>
<td>0.133</td>
<td>0.130</td>
</tr>
<tr>
<td>40%–50%</td>
<td>0.154</td>
<td>0.151</td>
</tr>
</tbody>
</table>

If there is a modest longitudinal decorrelation between the two subevents, it will manifest as a slight increase in the \( \delta_{2SE} \) parameter. The final \( v_n \) is averaged over that decorrelation. This effect, as in previous unfolding analyses \[10\], is neglected.

We highlight that even in the case of a perfect detector with perfect acceptance, there remains a smearing due to the finite particle number in each event. This raises a question regarding the meaning of a true \( v_n \) that is being unfolded back to. In a hydrodynamic description, there is a continuous fluid from which one can calculate a single true anisotropy \( v_n \) for each event. If one then has the fluid breakup into a finite number of particles \( N \), e.g. via Cooper-Frye freeze-out, the anisotropy of those
$N$ particles will fluctuate around the true fluid value. However, in a parton scattering description, for example AMPT [27], the time evolution is described in terms of a finite number of particles $N$. In this sense there is no separating of a true $v_n$ from that encoded in the $N$ particles themselves. Regardless, one can still mathematically apply the unfolding and compare experiment and theory as manipulated through the same algorithm.

As noted before, the one-dimensional radial projection of a two-dimensional Gaussian is the so-called Bessel-Gaussian function. In this case it means that the conditional probability to measure a value $v_n^{\text{obs}}$ given a true value $v_n$ has the following Bessel-Gaussian form:

$$
p(v_n^{\text{obs}}|v_n) \propto v_n^{\text{obs}} e^{\frac{-\left(v_n^{\text{obs}}-v_n\right)^2}{2\sigma^2}} I_0 \left( \frac{v_n^{\text{obs}}}{\sigma} \right), \tag{25}
$$

where $\sigma$ is the smearing parameter characterizing the response due to the finite particle number (including from the detector efficiency and acceptance), and $I_0$ is a modified Bessel function of the first kind. The smearing parameter $\sigma$ uses the combination of the two FVTX arms and is related to the result from the difference by $\sigma = \delta_{2SE}/2$. We highlight that the Bessel-Gaussian in Eqn. 25 is different from the Bessel-Gaussian in Eqn. 24, we can simply evaluate a large grid of parameter combinations as guesses for the truth distribution and forward fold them, i.e. passing them through the response matrix to compare to the observed $Q_n$ distribution. We have carried out such a “forward fold” procedure with over 10,000 parameter combinations. We then determine the statistical best fit parameters and their statistical uncertainties based on a $\chi^2$ mapping. We consider only the Gaussian statistical uncertainties here, and detail our treatment of systematic uncertainties in Section IV B. Some advantages of this procedure are that we explore the full $\chi^2$ space and have no sensitivity to an unfolding prior, regularization scheme, and number of iterations. The disadvantage of course is the ansatz that the distribution is precisely Bessel-Gaussian.

The Au+Au 20%–30% centrality class is expected to provide the best conditions, in terms of the predicted $\langle v_2 \rangle$ and resolution $\sigma$, to determine $p(v_2)$ via unfolding. However, because it is quite challenging to unfold the measured distribution directly, we constructed a test version of the problem to illustrate the procedure using the SVD method. The details of this test are given in Appendix V, but the end result is that the unfolding procedure is inherently unstable and therefore fails to converge for the resolution parameters in the present analysis.

As a result, instead of inverting the response matrix, we can make an ansatz that the probability distributions, $p(v_2)$ and $p(v_3)$ are exactly Bessel-Gaussian in form. Under this restrictive assumption, because the Bessel-Gaussian form has only two parameters as shown in Eqn. 24 we can simply evaluate a large grid of parameter combinations as guesses for the truth distribution and forward fold them, i.e. passing them through the response matrix to compare to the observed $Q_n$ distribution. We have carried out such a “forward fold” procedure with over 10,000 parameter combinations. We then determine the statistical best fit parameters and their statistical uncertainties based on a $\chi^2$ mapping. We consider only the Gaussian statistical uncertainties here, and detail our treatment of systematic uncertainties in Section IV B. Some advantages of this procedure are that we explore the full $\chi^2$ space and have no sensitivity to an unfolding prior, regularization scheme, and number of iterations. The disadvantage of course is the ansatz that the distribution is precisely Bessel-Gaussian.

Examples of this $\chi^2$ forward fold mapping are shown in Figure IV. It is striking that for the $v_2$ case in the Au+Au 20%–30% central bin, the forward folding reveals a tight constraint on the Bessel-Gaussian parameters. In contrast, for the same centrality bin and $v_3$, there is a band of parameter combinations providing a roughly equally good match to the experimental $Q_3$ distribution. Shown
FIG. 4. Example distribution of $Q$ for the (a), (b) elliptic $n = 2$ case and (c), (d) triangular $n = 3$ case (lower). (a) and (c) show the 2-dimensional distribution. (b) and (d) show the 1-dimensional distribution $|Q|$. The distribution corresponds to Au+Au collisions at $\sqrt{s_{NN}} = 200$ GeV and centrality 20–30%. Also shown in (b), (d) is the best fit for this run Bessel-Gaussian truth distribution and the corresponding forward folded result (i.e. pushing the truth distribution through the response matrix).

in Figure 4 for $v_2$ (upper) and $v_3$ (lower) are the best Bessel-Gaussian fit distributions (red) and their forward folded results (blue). Both cases show good agreement between the forward folded results and the measured experimental distribution. The corresponding best $\chi^2_{\text{min}}$ values indicate a good match. It is notable that in the $v_3$ case, the $\chi^2_{\text{min}}$ values are slightly worse in all cases and significantly worse in the most peripheral selection. The more peripheral data have a slightly larger tail at high $Q_3$ values which could indicate an incompatibility with the Bessel-Gaussian ansatz.
PHENIX Au+Au $\sqrt{s_{NN}}=200$ GeV 20-30\%

FIG. 5. Two dimensional color plot of $\chi^2 - \chi^2_{min}$ values as a function of the two Bessel-Gaussian parameters, $v^n_{RP}$ and $\delta_{vn}$ for the 20\%-30\% centrality bin. (a) shows the second harmonic and (b) shows the third harmonic. Only $\chi^2$ values up to $\chi^2_{min} + 25$ are shown.

IV. RESULTS AND DISCUSSION

Here we detail the full set of results for the elliptic and triangular flow moments and distributions in Au+Au collisions at $\sqrt{s_{NN}} = 200$ GeV. We start by detailing the cumulant results.

A. Cumulants Results

First we show in Figure 6 (a) and (b) the centrality dependence of $v_2$ and $v_4$, respectively. The statistical uncertainties are shown as vertical lines and the systematic uncertainty is quoted as a global factor un-}

\begin{align}
\chi_2 &= \frac{1}{N} \sum_i f_i^2 (\Delta f_i)^2 \\
\chi_4 &= \frac{1}{N} \sum_i f_i^2 (\Delta f_i)^4 \\
\chi_6 &= \frac{1}{N} \sum_i f_i^2 (\Delta f_i)^6
\end{align}

In addition, we present the $v_2$, $v_4$, and $v_6$ values for events with $2 < |\eta| < 3$, $p_T > 0.3$ GeV/c. These results are in good agreement with STAR [12] and PHOBOS [13].

Figure 7 shows the centrality dependence of multi-particle $v_2$, with 2, 4, 6, and 8 particles. The 4-, 6-, and 8-particle $v_2$ values are consistent with each other, as expected from the small-variance limit [7]. When accounting for the $\eta$-dependence of $v_2$ as measured by PHOBOS [32], which indicates that $v_2$ at 1 < $|\eta|$ < 3 is about 1.25 times lower than it is at $|\eta| < 1$, the 2-, 4-, and 6-particle cumulant $v_2$ are in good agreement with the STAR results [12].

Considering that $v_2 \approx \sqrt{v_2^2 + \sigma_{v_2}^2}$ and that in the small variance limit $v_4 \approx \sigma_{v_2}$, one can estimate the relative fluctuations as

$$\frac{\sigma_{v_2}}{v_2} \approx \sqrt{\frac{(v_2^2 - v_4^2)^2}{(v_2^2)^2 + (v_4^2)^2}}.$$
FIG. 7. Multi-particle $v_2$ as a function of centrality in Au+Au collisions at $\sqrt{s_{NN}} = 200$ GeV. The magenta open diamonds indicate the $v_2\{SP\}$, the blue open squares indicate $v_2\{4\}$, the black open circles indicate $v_2\{6\}$, and the green filled diamonds indicate $v_2\{8\}$.

eccentricity $\varepsilon_n$ distributions. If there is a linear mapping between initial spatial eccentricity and final momentum anisotropy ($\varepsilon_n \propto v_n$), we should expect a good match between $\sigma_{\varepsilon_n}/\langle \varepsilon_n \rangle$ and $\sigma_{v_n}/\langle v_n \rangle$. Also show in Fig. 8 is the Monte Carlo Glauber result via the calculation of cumulants (solid blue line), as well as the direct calculation of the variance and mean from the full $\varepsilon_n$ distribution (dashed blue line). One sees that in midcentral 10%–50% collisions, the data and both theory curves agree reasonably. For more central collisions, the Monte Carlo Glauber data-style calculation shows the same trend as the data whereas the Monte Carlo Glauber direct calculation is significantly lower. This is due to the fact that the small-variance limit is not a valid approximation in central collisions. In peripheral collisions, both Monte Carlo Glauber curves under-predict the data. This has been attributed to the nonlinear response in hydrodynamics [33].

FIG. 8. Cumulant method estimate of $\sigma_{v_2}/\langle v_2 \rangle$ as a function of centrality in Au+Au collisions at $\sqrt{s_{NN}} = 200$ GeV. The data are shown as black open squares. The same calculation as done in data is done in AMPT, shown as a solid green line. Calculations of $\sigma_{v_2}/\langle v_2 \rangle$ performed in the Monte Carlo Glauber model are shown as blue lines. The solid blue line is the Monte Carlo Glauber calculation done using the same estimate as the data, the dashed blue line is the direct calculation of the moments of the MC Glauber $\varepsilon_2$ distribution.

Now we consider the $v_3$ case. Figure 9 shows $v_3\{2, |\Delta\eta| > 2\}$ as a function of centrality. The centrality dependence of $v_3$ is much smaller than that of $v_2$. The calculations for this figure were performed with the same centrality dependence in centrality as that of $v_2$. The data are shown as black open squares. The same calculation as done in data is done in AMPT, shown as a solid green line. Calculations of $\sigma_{v_3}/\langle v_3 \rangle$ performed in the Monte Carlo Glauber model are shown as blue lines. The solid blue line is the Monte Carlo Glauber calculation done using the same estimate as the data, the dashed blue line is the direct calculation of the moments of the MC Glauber $\varepsilon_2$ distribution.
which is expected because triangular flow is generated dominantly through fluctuations.

![Graph](image)

**FIG. 9.** Centrality dependence of \(v_3\{2, |\Delta \eta| > 2\}\) in Au+Au collisions at \(\sqrt{s_{NN}} = 200\) GeV, shown as magenta diamonds. The systematic uncertainty is indicated by a shaded magenta band. Also shown as black squares are \(\sqrt{v_3^2}\) as determined from the folding analysis, which is shown in the next section.

Figure 10 (a) shows the results for \(c_3\{4\}\) as a function of centrality. The results are always positive within the systematic uncertainties and shows a trend towards even larger positive values as one moves away from the most central collisions. Since \(v_3\{4\} = (c_3\{4\})^{1/4}\), the \(v_3\{4\}\) are complex-valued.

**B. Folding Results**

Now we turn to the results from the event-by-event forward fold. As detailed in Section III B in the \(v_2\) case the Bessel-Gaussian parameters are well-constrained apart from the most central events. In the \(v_3\) case, however, the Bessel-Gaussian parameters are not well-constrained for any centrality class. However, despite the broad range of possible \(\delta_{v_3}\) and \(v_3^\text{RP}\) values, these correspond to a rather small range for the real mean \(\langle v_3 \rangle\) and root-mean-square or variance \(\sigma_{v_3}\) of the distributions. This means that despite the lack of constraint on the parameters, the first \(\langle v_3 \rangle\) and second \(\sigma_{v_3}\) moments of the distribution are nevertheless well-constrained.

We can quantify \(\langle v_3 \rangle\) and \(\sigma_{v_3}\) by varying the Bessel-Gaussian parameters within the one- and two-standard deviation statistical constraints. In addition, we determine the systematic uncertainties on these quantities by varying the \(z\)-vertex and analyzing loose and tight cuts (as described for the cumulants analysis). An additional systematic uncertainty on the response matrix is estimated by splitting the data sample into two subsets, one with higher extracted \(\delta\) and one with lower, forward folding the two data sets separately, and then assessing the difference.

Figure 11 (a) shows the extracted first moment \(\langle v_2 \rangle\), Fig. 11 (b) shows the extracted second moment \(\sigma_{v_2}\), and Fig. 11 (c) shows the relative fluctuations \(\sigma_{v_2}/\langle v_2 \rangle\), each as determined from the folding method and as a function of centrality. Likewise, Fig. 12 (a) shows the extracted \(\langle v_3 \rangle\), Fig. 12 (b) shows the extracted \(\sigma_{v_3}\), and Fig. 12 (c) shows the relative fluctuations \(\sigma_{v_3}/\langle v_3 \rangle\). The colored bands indicate the statistical uncertainties at the 68.27% confidence level (red) and the 95.45% confidence.
consistent within the systematic uncertainties. These results are determined from the cumulant method as shown in Figure 8, except in the most central and peripheral Au+Au events. The most central 0%-5% events are consistent within the systematic uncertainties.

We highlight that the \( \sigma_{v_2}/\langle v_2 \rangle \) values agree well with those determined from the cumulant method as shown in Figure 8 except in the most central and peripheral Au+Au events. The most central 0%-5% events are exactly where the Monte Carlo Glauber results in Figure 8 indicate a breakdown in the small-variance approximation. This is a good validation of the forward folding procedure and another confirmation that the event-by-event elliptic flow fluctuations in Au+Au collisions at \( \sqrt{s_{NN}} = 200 \text{ GeV} \) are dominated by initial geometry fluctuations.

Intriguingly, whereas the values of \( \sigma_{v_2}/\langle v_2 \rangle \) vary significantly as a function of centrality, the values of \( \sigma_{v_3}/\langle v_3 \rangle \) are almost precisely 0.52 independent of centrality. To understand this better, we need to consider a rather peculiar feature of the Bessel-Gaussian Function. Figure 13 shows the \( \sigma_{v_3}/\langle v_3 \rangle \) of the Bessel-Gaussian as a function of the ratio \( \delta/\langle v_n^{RP} \rangle \). For values of \( \delta > \langle v_n^{RP} \rangle \), the observed \( \sigma_{v_3}/\langle v_3 \rangle \) saturates at a value of about 0.52. Thus, any Bessel-Gaussian in the large variance limit will have a \( \sigma_{v_3}/\langle v_3 \rangle \) of this same value.

This observation can, in fact, help shed light on the observed discrepancy between the CMS [11] and ATLAS [10] data on \( \sigma_{v_3}/\langle v_3 \rangle \). Figure 14 shows \( \sigma_{v_2}/\langle v_2 \rangle \) and \( \sigma_{v_3}/\langle v_3 \rangle \) as a function of centrality in Pb+Pb collisions at \( \sqrt{s_{NN}} = 2.76 \text{ TeV} \) from CMS and ATLAS. The CMS results are obtained using the cumulant method assuming the small-variance limit. In contrast the ATLAS results are obtained via an event-by-event unfolding and calculating the exact mean and variance of the distribution.

The \( \sigma_{v_2}/\langle v_2 \rangle \) values are in very good agreement, which appears to validate the small variance approximation (as was also validated in the Au+Au at \( \sqrt{s_{NN}} = 200 \text{ GeV} \) case in this analysis). In contrast, there is a large difference in the \( \sigma_{v_3}/\langle v_3 \rangle \) between the different methods. The ATLAS \( \sigma_{v_3}/\langle v_3 \rangle \) values are all very close to 0.52, exactly as observed above in the present Au+Au data and as found to be a limiting case for the Bessel-Gaussian function. To better understand the \( \sigma_{v_3}/\langle v_3 \rangle \), we also show \( \sigma_{v_3}/\langle v_3 \rangle \) as determined from MC Glauber calculations. The dashed red-line uses the small-variance limit estimate with cumulants, as is done for the CMS data, and the agreement is quite reasonable. The solid red line is calculated from the moments of the \( \varepsilon_3 \) distribution directly, and shows good agreement with the ATLAS data. This represents a quantitative confirmation of the event-by-event fluctuations and the breakdown in the small variance approximation. The \( v_3 \{4\} \) at forward rapidity at RHIC is found to be complex-valued, which may be the result of a very small flow \( v_3 \) and significant nonflow contributions.

V. SUMMARY AND CONCLUSIONS

In summary, we have presented measurements of elliptic and triangular flow in Au+Au collisions at 200 GeV for charged hadrons at forward rapidity 1 < \( |\eta| \) < 3. In particular, we compare flow cumulants \( v_2 \{2\} \), \( v_2 \{4\} \), \( v_2 \{6\} \), \( v_2 \{8\} \) and \( v_3 \{2\} \), \( v_3 \{4\} \) and the mean and variance of the \( v_2 \) and \( v_3 \) event-by-event distributions using
FIG. 11. Folding results for (a) $v_2$, (b) $\sigma_{v_2}$, and (c) $\sigma_{v_2}/\langle v_2 \rangle$. The black lines above and below the points indicate the systematic uncertainties. The red (green) boxes indicate the statistical uncertainties at the 68.27% (95.45%) confidence level. In the case of $\langle v_2 \rangle$, the statistical uncertainties at the 68.27% confidence level are too small to be seen, and the uncertainties at the 95.45% confidence level are visible but noticeably smaller than the marker size. Shown as blue squares are the same quantities as determined using the cumulant based calculation—these points are slightly offset in the $x$-coordinate to improve visibility.

FIG. 12. Folding results for (a) $\langle v_3 \rangle$, (b) $\sigma_{v_3}$, and (c) $\sigma_{v_3}/\langle v_3 \rangle$. The black lines above and below the points indicate the systematic uncertainties. The red (green) boxes indicate the statistical uncertainties at the 68.27% (95.45%) confidence level. The $\sigma_{v_3}/\langle v_3 \rangle$ values are all $\approx 0.52$, the apparent limiting value of this quantity for the Bessel-Gaussian distribution.

a forward-fold procedure with a Bessel-Gaussian ansatz. These measurements are complementary in terms of sensitivity to initial state geometry fluctuations and additional fluctuations from the evolution of the medium, for example via dissipative hydrodynamics.

In the small-variance limit, where the event-by-event flow fluctuations are small compared to the average flow value i.e. $\sigma_{v_n}/\langle v_n \rangle < 1$, we expect the cumulants extraction and the forward-fold results to agree. This is the case for elliptic flow in Au+Au collisions from 10%–50% central and both results agree with event-by-event fluctuations in the initial geometry as calculated via Monte Carlo Glauber.

In contrast, we find that the small-variance limit fails for triangular flow for all centralities at RHIC and the LHC. For LHC Pb+Pb results, the large-variance result for the cumulants can be described purely via Monte Carlo Glauber initial geometry fluctuations. However, for RHIC Au+Au collisions the complex values of $v_3\{4\}$ indicate that there may be additional nonflow influences as well as sources of fluctuations in the translation of initial geometry into final state momentum triangular anisotropies. Detailed comparisons with event-by-event hydrodynamic calculations should be elucidating to understand the nature of these fluctuations.
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APPENDIX: TEST CASE FOR FULL UNFOLD

For this test case, the response matrix $\hat{A}$ is shown in Fig. 15 (a) and is identical to that for the real data 20%–30% centrality class. We then attempt to solve the inverse problem $\hat{A}\hat{Q}_{2}\text{obs} = \hat{Q}_{2}\text{obs}$, where $\hat{Q}_{2}\text{obs}$ has been obtained in the limit of infinite statistical precision, assuming a truth-level distribution with parameters such that the smearing, as encoded in $\hat{A}$, yields a distribution similar to that measured in data. The singular value factorization $\hat{A} = U\Sigma W^T$ of the matrix is obtained, where $U$ and $W$ are unitary matrices whose column vectors, $u_i$ and $w_i$, are the left- and right- singular vectors of $\hat{A}$, respectively, and $\Sigma$ is a diagonal matrix, whose nonzero entries $\sigma_i$ are its singular values. Figure 15 (b) shows a few selected right-singular vectors $w_i$. Notice that some vectors, namely those corresponding to the largest singular values, are harmonic, whereas those corresponding to the smallest singular values are essentially noise.

Because the response matrix is singular, we use the SVD decomposition to construct the solution of the inverse problem as a linear combination of all right-singular
vectors, as follows:

\[ \tilde{Q}_2 = \sum_{i=1}^{\text{Dim}(A)} \varphi_i \left( \frac{\tilde{u}_i^T \cdot \tilde{Q}^{\text{obs}}_{2 \lambda}}{\sigma_i} \right) \tilde{u}_i. \]  

(27)

The damping factors \( \varphi_i = \sigma_i^2/(\sigma_i^2 + \lambda^2) \), for some \( \lambda \in \mathbb{R} \), are introduced to attenuate the contribution of the noisy singular vectors to the sum. It is important to point out that in most implementations of SVD used in high-energy physics, including RooUnfold, the above sum is simply truncated to include only a subset of the harmonic singular vectors, potentially leading to loss of information.

To determine which singular vectors contribute to the solution in a meaningful manner, it is useful to examine the Picard plot for the problem at hand, shown in Fig. 15 (c), which displays the singular values \( \sigma_i \) of \( \tilde{A} \), as well as the projection of \( Q^{\text{obs}}_2 \) onto the singular vectors \( \tilde{u}_i^T \cdot \tilde{Q}^{\text{obs}}_2 \), and the solution coefficients \( \tilde{u}_i^T \cdot \tilde{Q}^{\text{obs}}_2 / \sigma_i \). Notice that the singular values and the Fourier coefficients drop sharply many orders of magnitude before leveling off, yet in such a way that their ratio is roughly constant. The implication is then that all singular vectors appear to contribute equally to the solution, which is clearly problematic given the noisy nature of most of them. In general, it is desirable for Fourier coefficients to drop off faster than the singular values (to fulfill the so-called discrete Picard condition), such that the Picard plot will reveal the appropriate set of terms to include in the solution, as identified by a sharp drop in the solution coefficients.

Given that our problem does not satisfy the Picard condition, we introduce the attenuation factors \( \varphi_i \) in Eqn. 27. The resulting unfolded \( Q_2 \) is shown in Fig. 15 (d), along with the true \( Q^{\text{true}}_2 \) and smeared \( Q^{\text{obs}}_2 \). We observe that the unfolding works well, yielding a good description of the true distribution shape, with uncertainties associated with varying the regularization parameter \( \lambda \).

However, in this case the unfolding procedure constitutes an ill-posed inverse problem, such that small perturbations in the input vector—that is, \( Q^{\text{obs}}_2 \)—translate to very large errors in the solution, compounded by the fact that the Picard condition is violated. In particular, we have verified with our test problem that the statistical fluctuations in \( Q^{\text{obs}}_2 \) when sampling a finite number of events, comparable to those recorded in data, indeed limit the number of available harmonic singular vectors, thus causing the solution to be dominated by noise.

We now examine the application of the above unfolding method to data. Fig. 16 (a) shows an ansatz for \( Q^{\text{true}}_2 \) assuming a Bessel-Gaussian form, and the corresponding refolded smeared distribution. It compares very well to the data, as shown in the ratio plot in Fig. 16 (b). In principle, given the good quality of the fit, one would expect the unfolding procedure to work with the data as input. However, the statistical fluctuations apparent in the ratio plot perturb the solution in such a way that the noisy nonharmonic singular vectors are enhanced even more than in the test problem, as shown in Fig. 16 (c). As a result, the number of available harmonic singular vectors is reduced, and the problem has no satisfactory solution, even when regularization is applied. Thus, to be explicit, the unfolding procedure fails. We note that if we apply our test example with a significantly better resolution, i.e. as in the ATLAS Pb+Pb case, the method does converge as expected.

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FIG. 15. (a) Response matrix for test unfolding problem; (b) Selected right-singular values of the response matrix; (c) Picard plot for inverse problem $\hat{A}Q_{\text{true}}^2 = Q_{\text{obs}}^2$, see text for details; (d) True, smeared, and unfolded $Q_2$ as determined using SVD with attenuation factors.

FIG. 16. (a) Data distribution for $Q_{\text{meas}}^2$ in Au+Au collisions at $\sqrt{s_{NN}} = 200$ GeV in the 20%–30% centrality class. Also shown is an assumed Bessel-Gaussian truth distribution and its resolution smeared result. (b) Data divided by the resolution smeared solution showing a good agreement within statistical uncertainties. (c) Picard plot for inverse problem with data $\hat{A}Q_{\text{true}}^2 = Q_{\text{obs}}^2$, see text for details;